

Figure 2

Examples of Program Information

Title = Seinfield
Program Type = Sitcom
Category = Comedy
Actors = (Actor1, Actor2)

Title = US Debt Report

Program Type = News article

Category = US Govt. Financial

People Mentioned = (Bill Clinton,

Alan Greenspan)

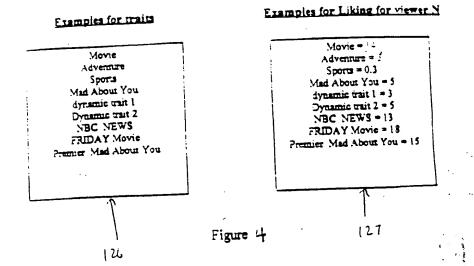
Example 1

Example 2

125

124

Figure 3



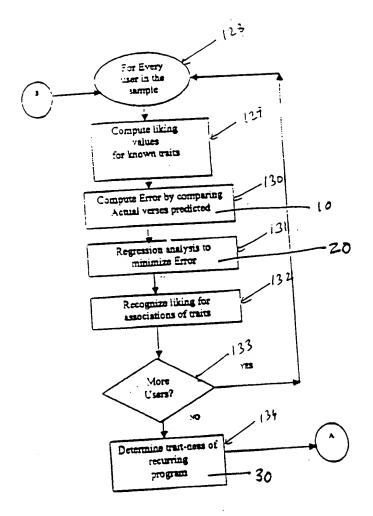


Figure 5(a)

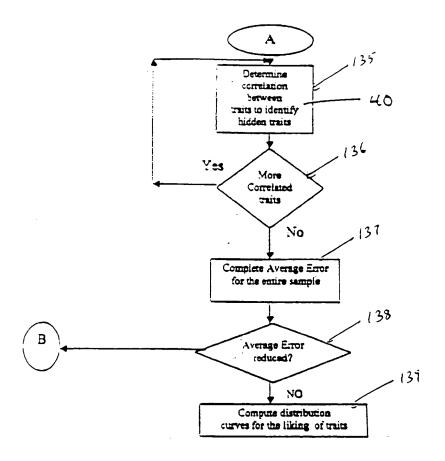


Figure 5 (b)

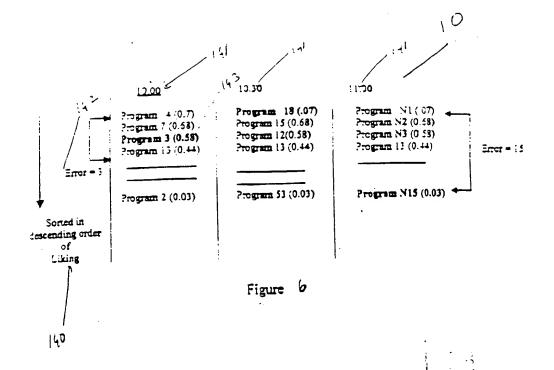


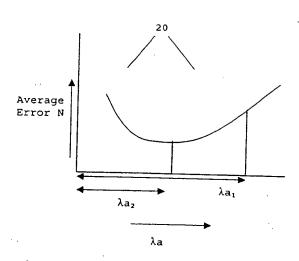
Figure 7

Current Liking Value

 $\lambda a_1 = 2$ $\lambda b_1 = 5$ $\lambda c_1 = -3$ $\lambda d_1 = 0$

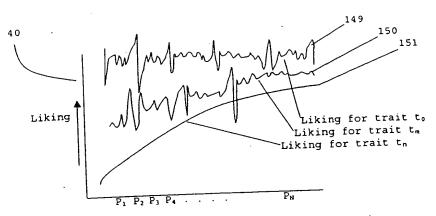
Next Liking Value

 $\lambda a_2 = 1.5$ $\lambda a_1 = 5$ $\lambda a_1 = -3$ $\lambda a_1 = 0$



 $(\lambda b = \lambda b_1$ $\lambda c = \lambda c_1$ $\lambda d = \lambda d_1$.

Figure 8



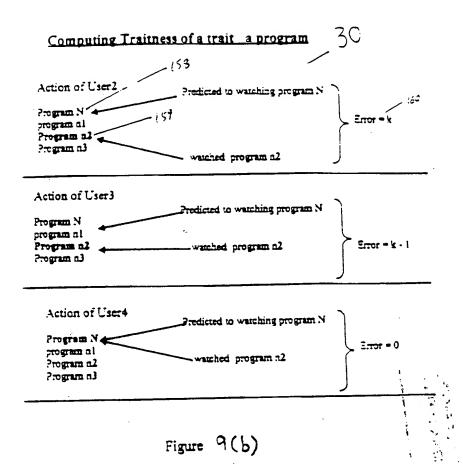
 t_{m} and t_{n} are correlated

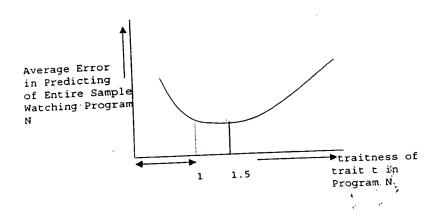
 t_m can be expressed as $t_m = t_x + t_m'$ t_n can be expressed as $t_n = a_x t_x + t_n'$

and

*

Figure 9(a)





Optimal value of traitness

e.g. comedy-ness in Seinfeld = 1.5 comedy-ness in Frasier = 0.89

Example for Liking Distribution Record format

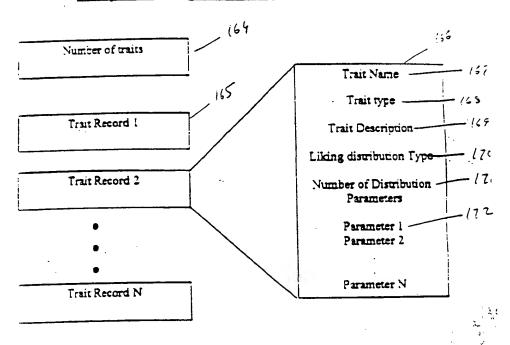


Figure 10

Some Sample Values For Fields in trait Record

Trait type

Static , dynamic Association Generated

Trait Description

(NBC . "NEWS"), SUBSTRING("CLA") IN DESC. TITLE

Distribution

Normal Exponential Defined type 1 Defined type 2

Distribution Parameters

Mean = 13, Deviation =2

Figure 11

Example for Traitness of recurring Programs

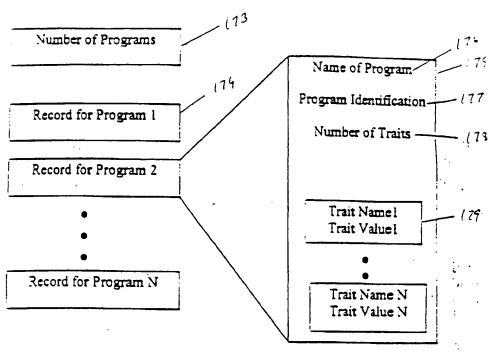


Figure !2

Example For Broadcasting traitness as a part of EPG Data

Program Info Seinfield, NBC, Comedy = 0.07 sitcom, Dynamic trait 1 = 0.1

Actor = Seinfield

Figure 13

Example for Selection Record

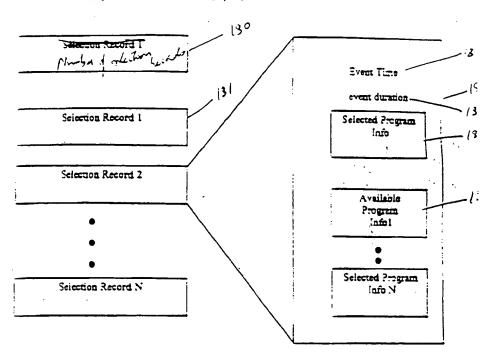


Figure 14

Generation of User Selection History

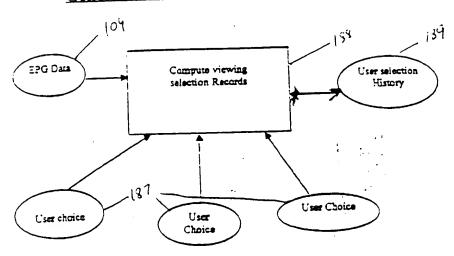


Figure 15

Learning Liking for traits for a given user

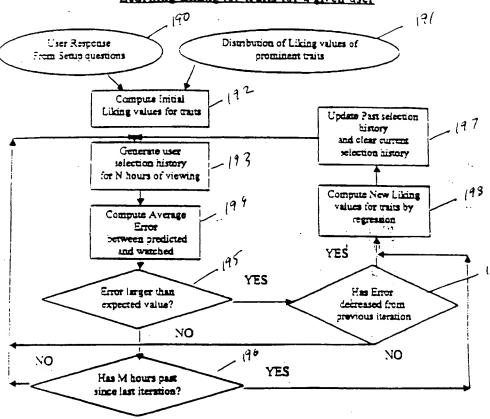


Figure 16

Computing Relevance

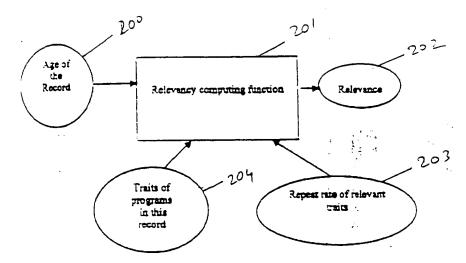


Figure 17 (a)

Figure 17(b)

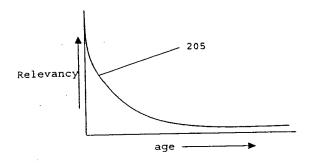
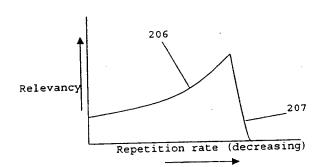


Figure 17(c)



Updation of past History

. 4

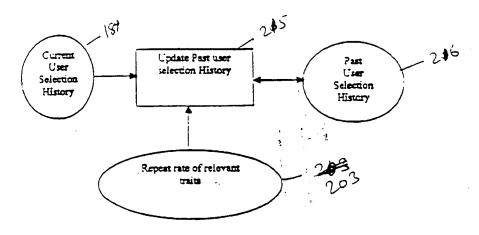
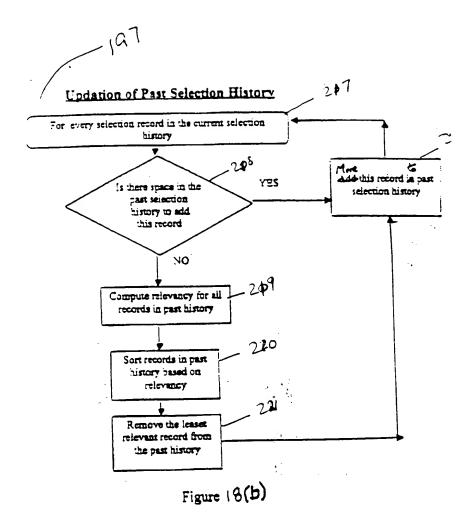
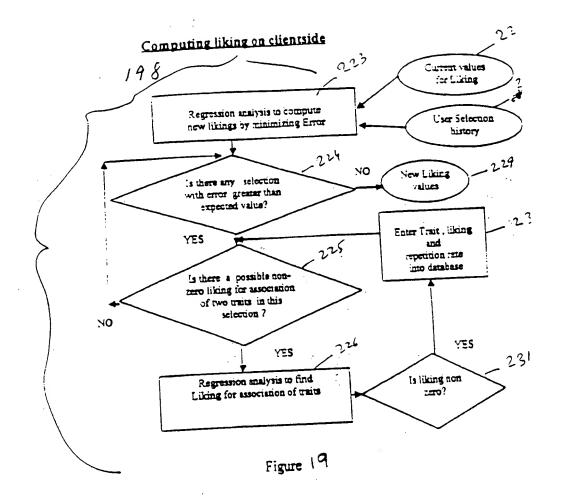


Figure 18 (a)





Computing scores for programs for future prediction

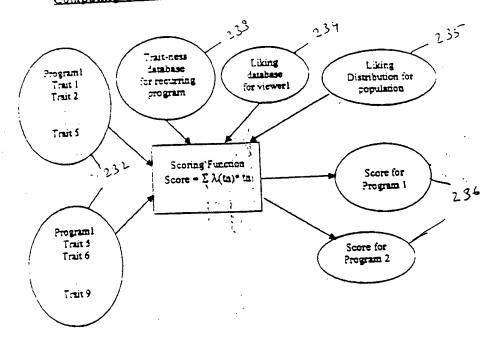
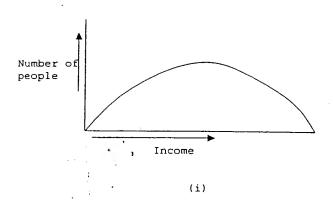
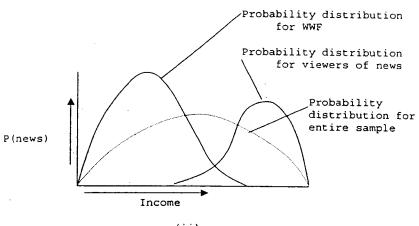


Figure 20

Figure 21(a)





(ii)

Enter Sunte

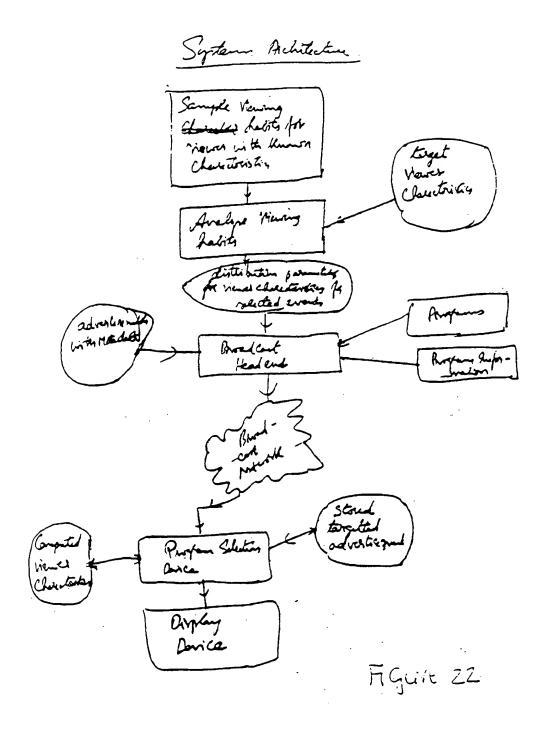
FRIVILLE T

Male Female

fort on and

Female Male

Figure 21 b



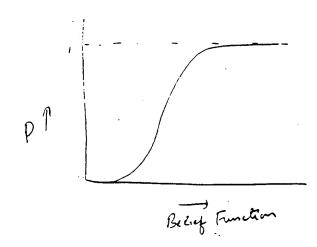


Figure 239

Demographic Trait Record format

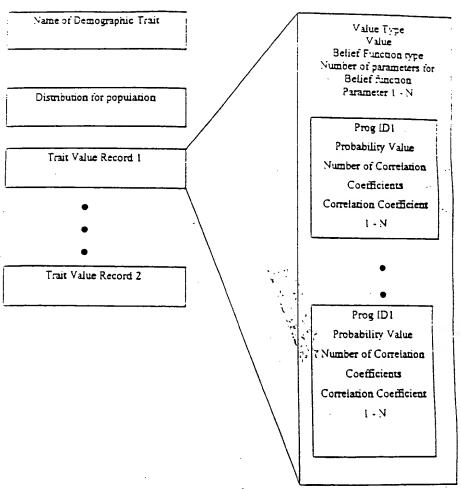


Figure 23b

Advertisement Targeting Record format

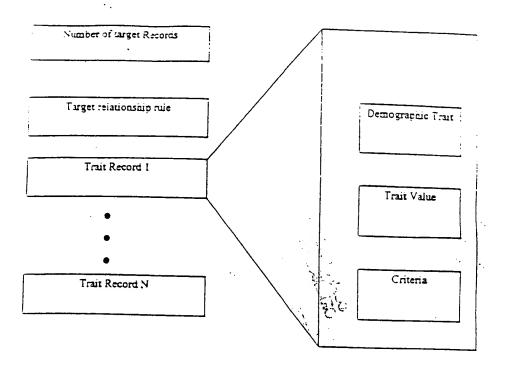
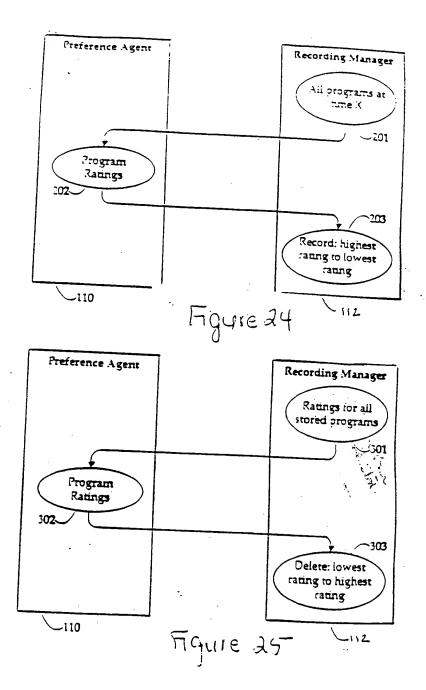
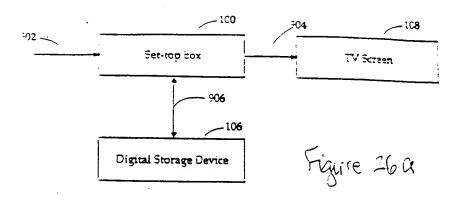


Figure 236





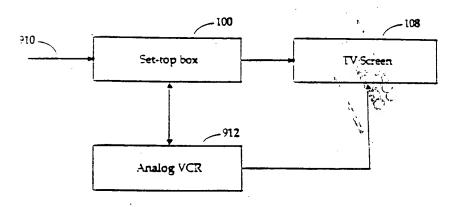


Figure 26b

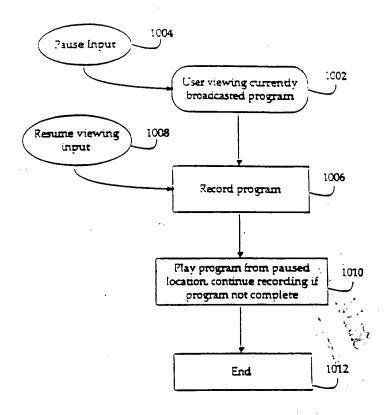


Figure 27

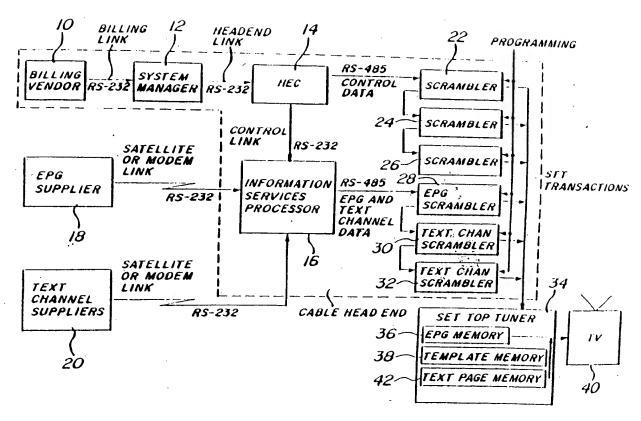


Figure 28

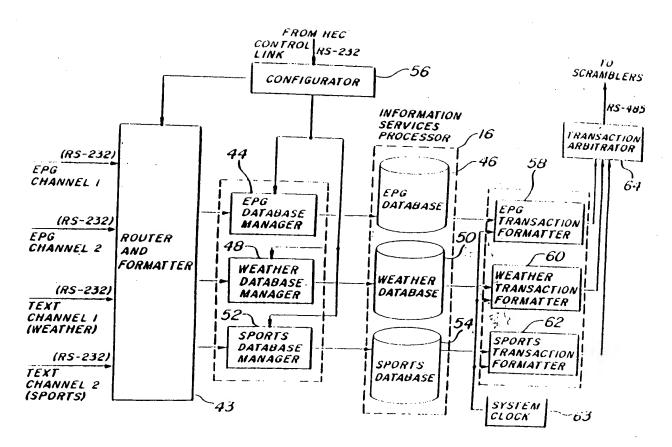


Figure 29

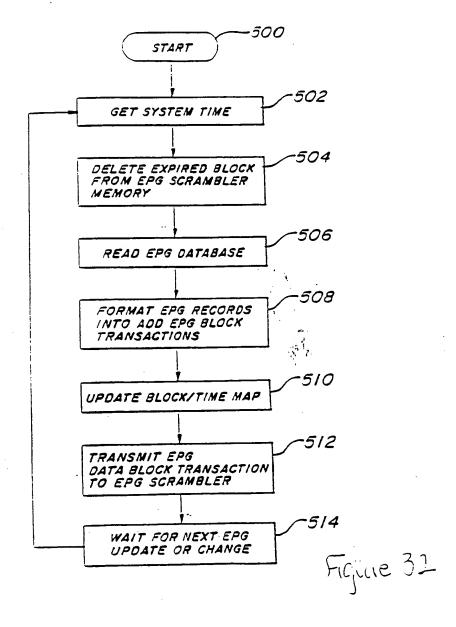
(INFORMATION FIELD)

| OATE AND CHANNEL OURATION RE | PEAT RATIN | G CATESORY | |
|------------------------------|------------|------------|---|
| RECORD KEY | | | |
| CRITIQUE ATTRIBUTES | TRAITS | TEXT DATA | |
| | COMPR | | 7 |
| | | Figure | 3 |

| SEGINNING STATION CONTROL FLAG ADDRESS 18YTE 18YTE 18YTE | INFOR- MATION FIELD A BYTES | FRAME E CHECK 2 BYTES | NDING |
|--|--------------------------------------|-----------------------------|-------|
|--|--------------------------------------|-----------------------------|-------|

Figure 31

EPG TRANSACTION FORMATTER 58



TEXT CHANNEL TRANSACTION FORMATTER 60,62

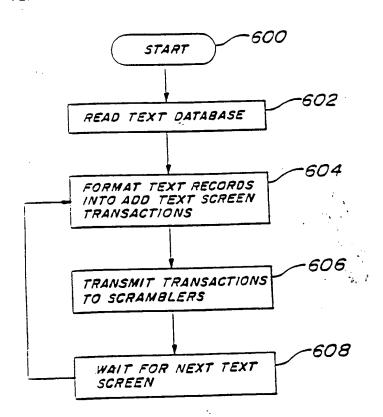
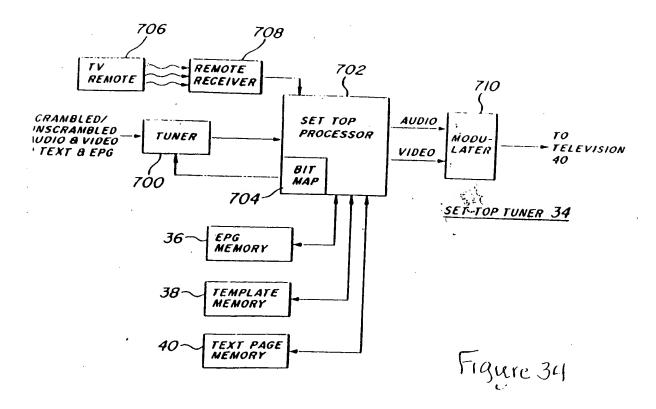


Figure 33



Process for automatically desating multiple profiles and suffermatically regardlying durantity agine profiles

| | Determine the number of usage profiles for the device |
|---|---|
| | |
| | Y |
| | Monitor user actions |
| • | |
| • | C |
| | Generate a history of user. |
| | 3C:IOU2 |
| | |
| | |
| | Ψ |
| | Generate multiple profiles |
| | |
| | |
| | · · |
| | Monitor Current user actions |
| | actions |
| | |
| | |
| | ▼ |
| | Predict the profiles that are |
| | |

Figure 35

CIRCULAR PROGRAM GUIDE

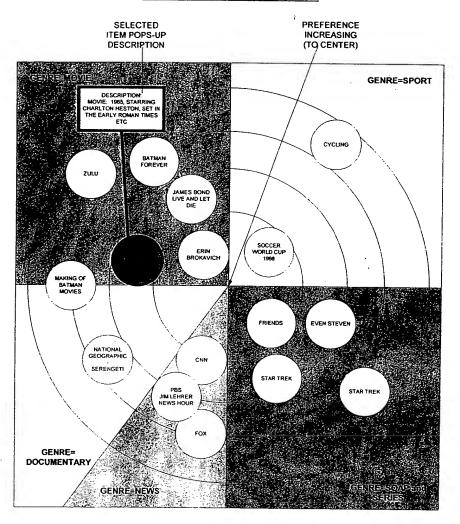


Figure 36